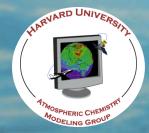
#### **Harvard University**

# Estimating US methane emissions using GOSAT observations

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## Why do we care about atmospheric methane?

- 1) Methane is a potent greenhouse gas
  - ▶ 2<sup>nd</sup> only to CO<sub>2</sub>

Emissions based radiative forcing [W m<sup>-2</sup>]

O.O O.S 1.O 1.5

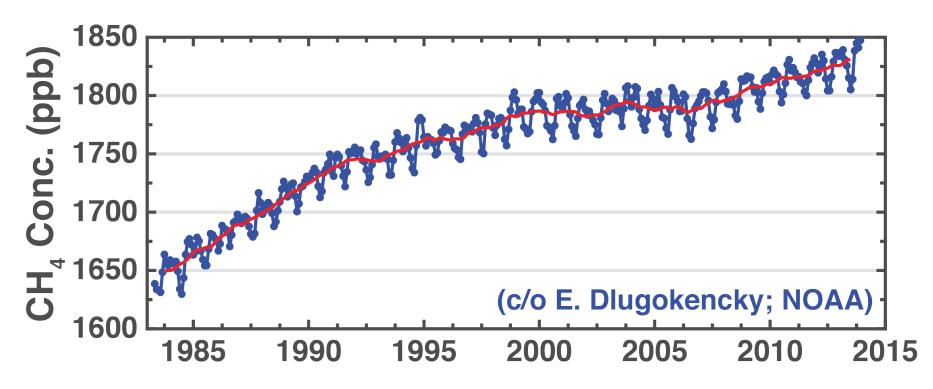
CO<sub>2</sub>

CH<sub>4</sub>

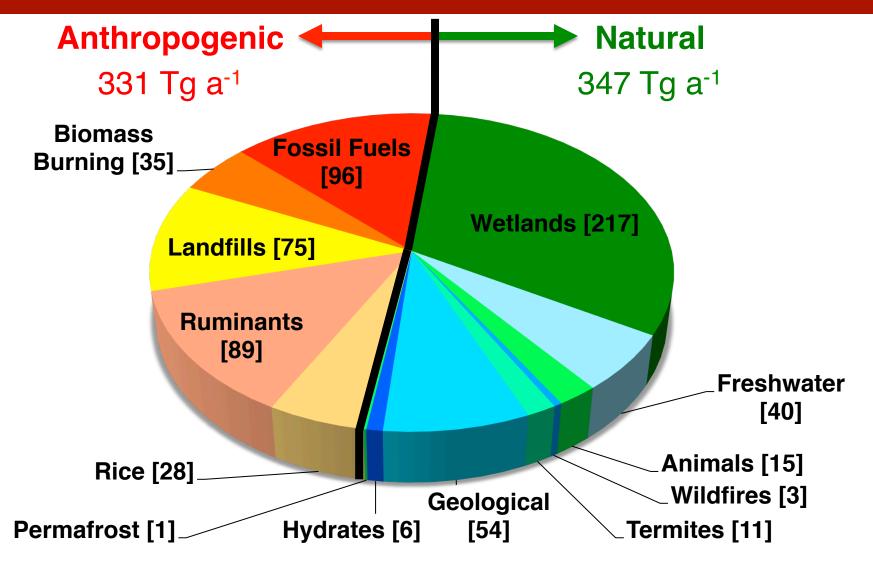
O<sub>3</sub>

IPCC (2013) H<sub>2</sub>O (strat)

2) Recent trends in atmospheric methane are not well understood



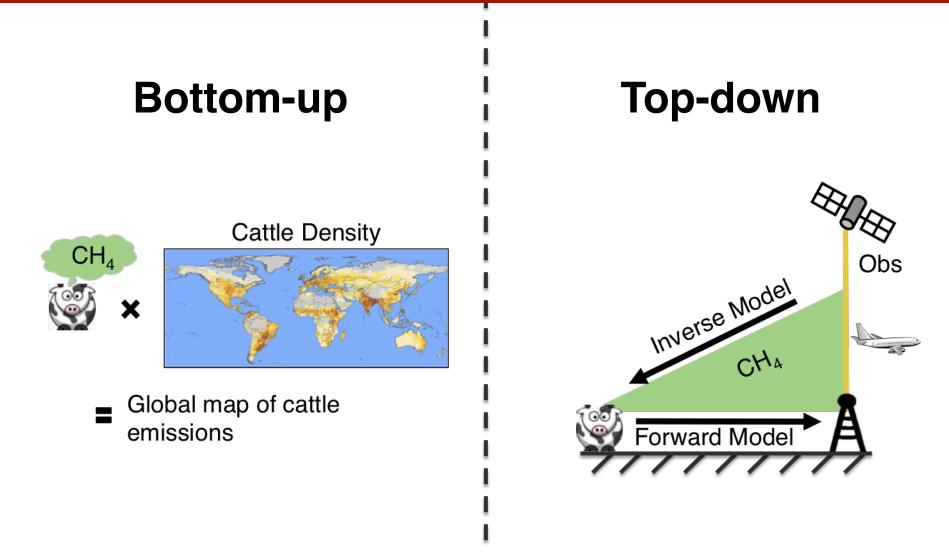
#### Global methane emission sources



**Biogenic Thermogenic** 

**Pyrogenic** 

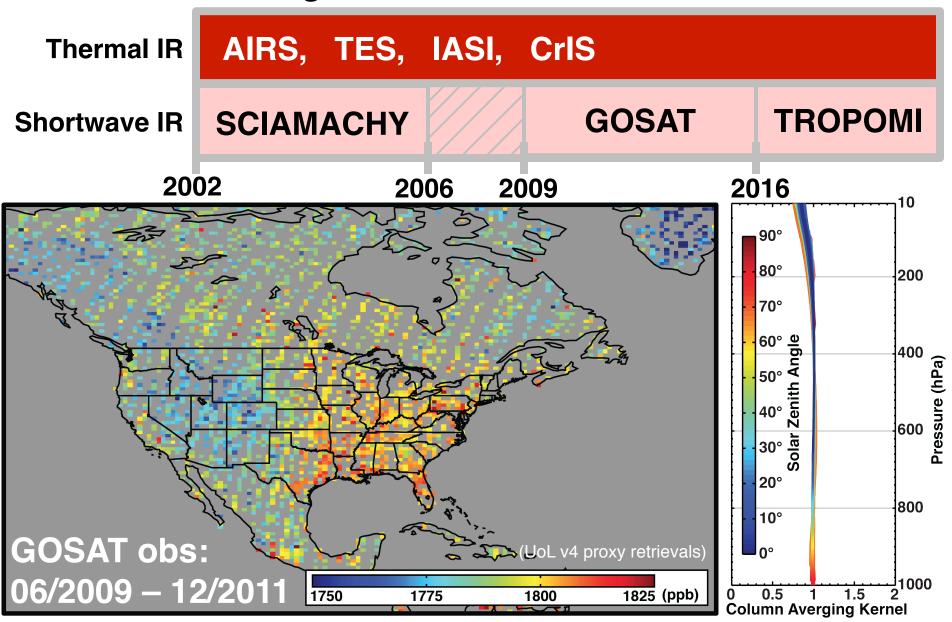
## Methods of estimating methane emissions



Satellites provide dense spatial coverage but have large uncertainties

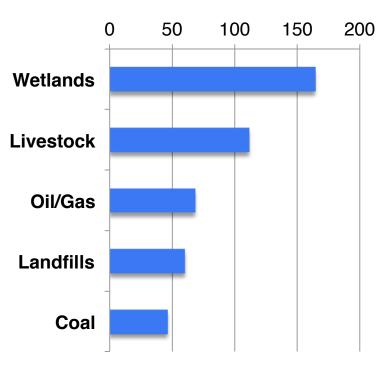
#### Retrievals of methane from observed radiances

#### **Satellites Observing Methane**

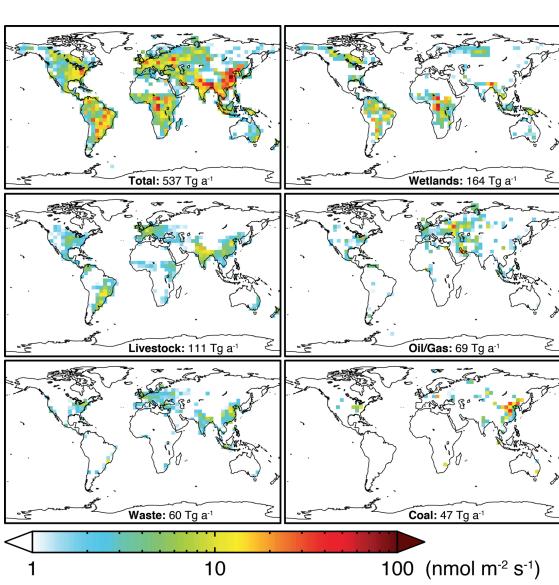


#### Prior emissions from EDGARv4.2 + LPJ + GFED3

## Major Sources (Tg yr<sup>-1</sup>)

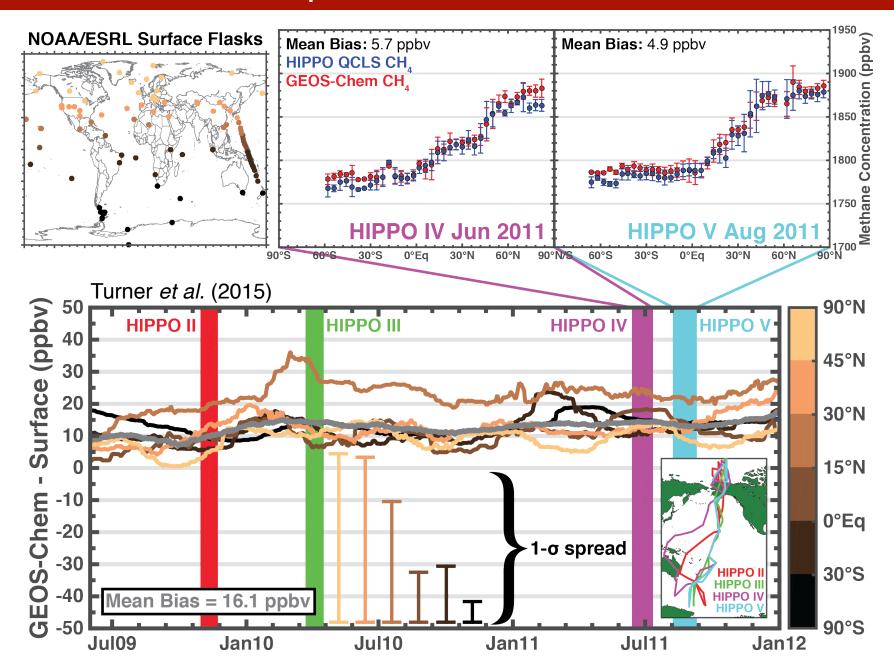


Total: 537 Tg yr<sup>-1</sup>

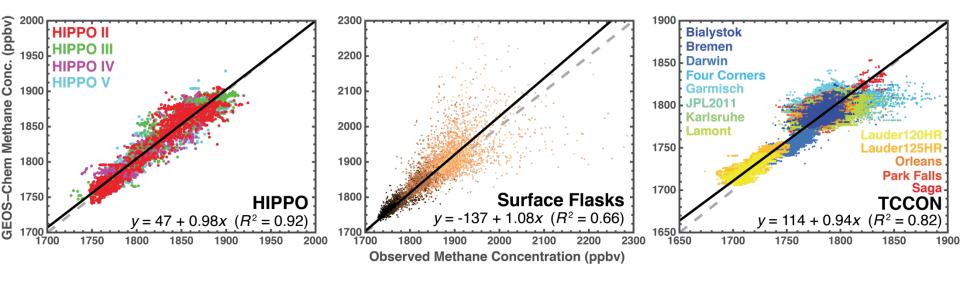


Turner *et al.* (2015)

## Model compares well with observations

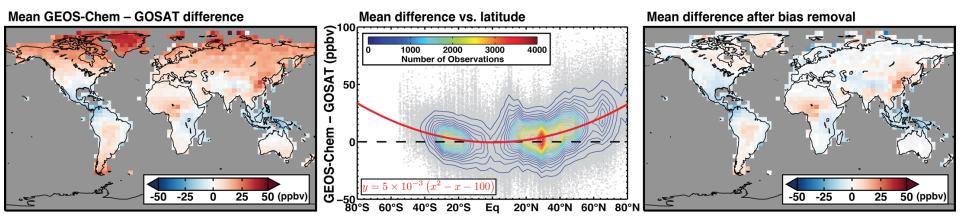


## Model compares well with observations



- Latitudinal gradient and seasonal cycle are represented
  - ► Compared to HIPPO, NOAA/ESRL, and TCCON
- Captures surface, free trop, and total column background

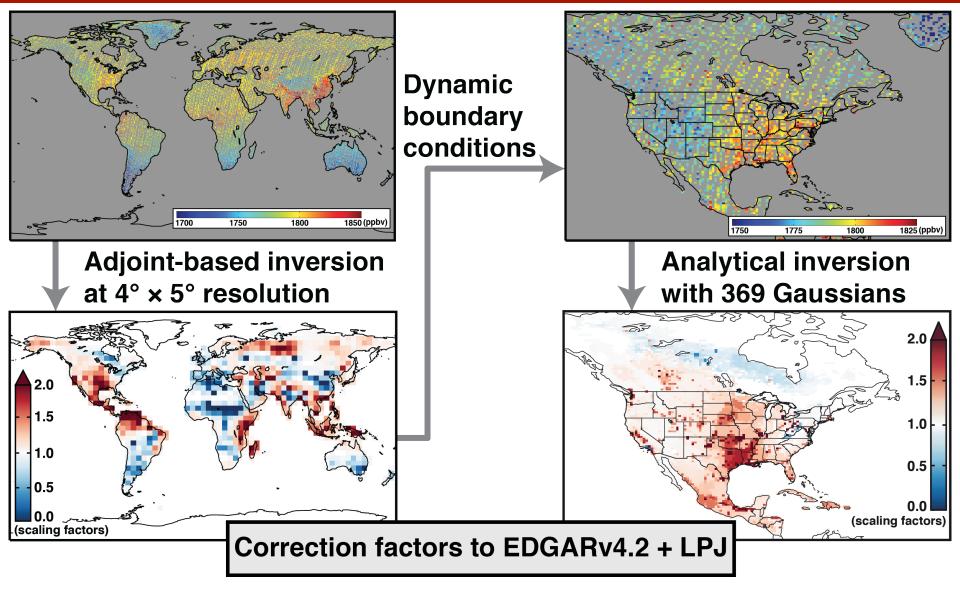
## Identifying a GOSAT/GEOS-Chem bias



- Model/satellite comparison identifies a high-latitude bias
  - Latitudinal bias not seen in surface, aircraft, or column comparison
- ▶ Remove bias before estimating methane emissions
  - Bias is either due to the model stratosphere or GOSAT retrievals

# Observations are ready for inversion!

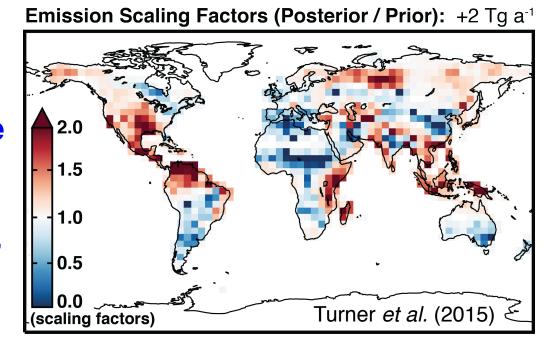
#### General inversion framework: 2009–2011 GOSAT data

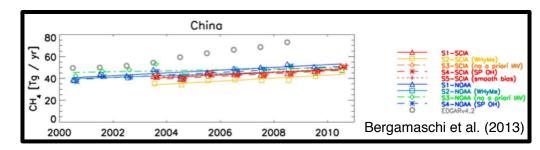


Global inversion provides dynamic BCs for North America

#### Global inversion results

- Overestimate of Chinese methane emissions
  - Consistent with previous work (e.g., Bergamaschi et al. 2013, Bruhwiler et al. 2014, Schwietzke et al. 2014)

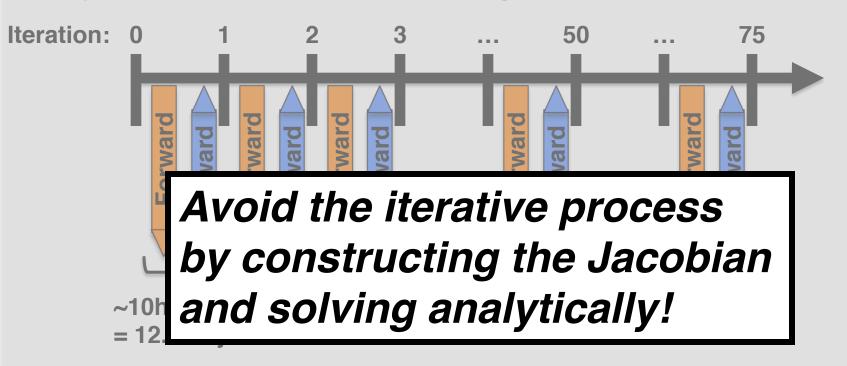




- Underestimate in South-Central US emissions
  - Will further investigate using Nested North American simulation

# Estimating methane emissions at high resolution

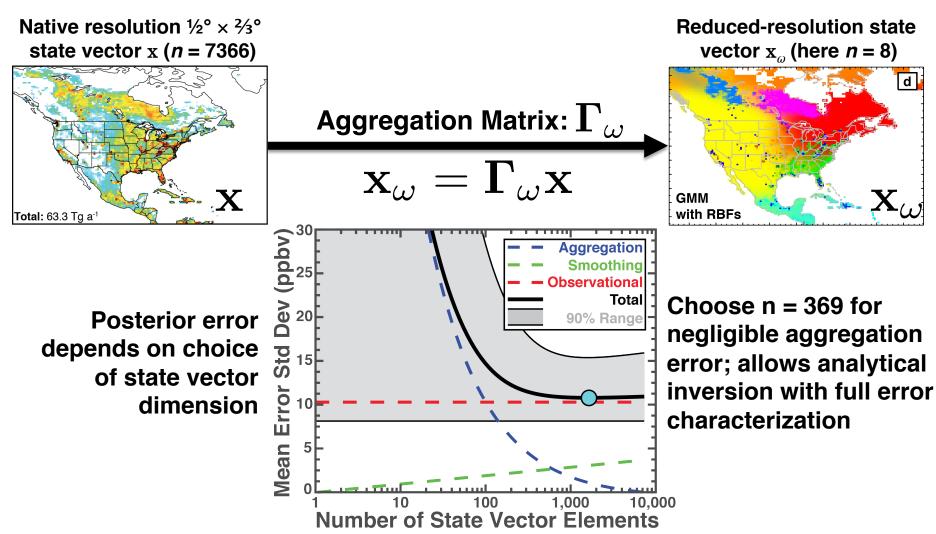
Adjoint is not ideal for long time horizons at hi-res



Simulation Walltime: 2.6 years

# Estimating methane emissions at high resolution

#### Spatial error correlations are important at fine spatial scales!

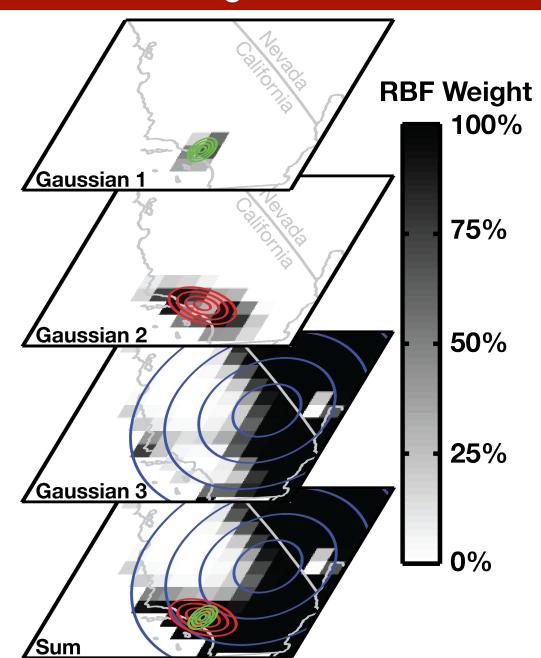


Optimal size must balance aggregation and smoothing error

## Radial Basis Functions retain high resolution

- Decompose the state vector into Gaussians
  - Group based on correlated prior emission patterns

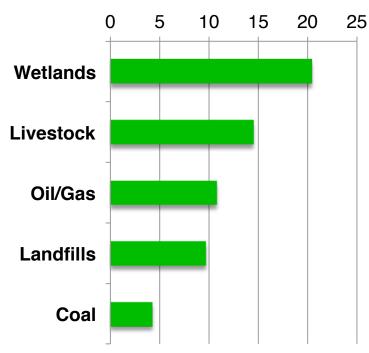
- Retain high resolution
  - Coarsen weak or uniform signals



Turner and Jacob (2015)

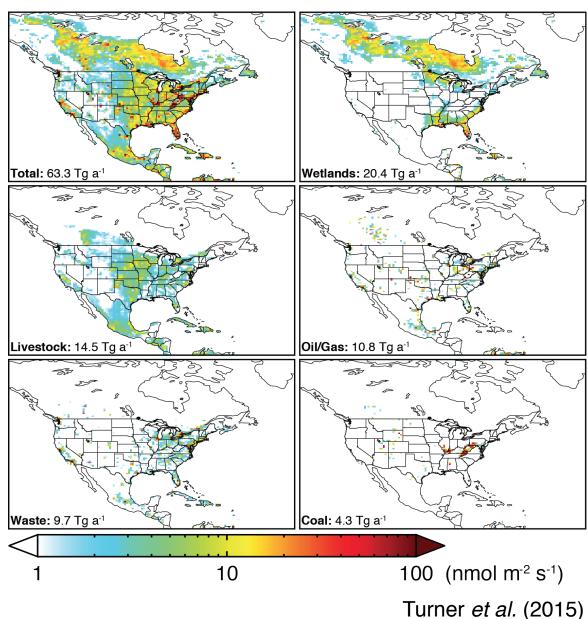
#### Prior methane emissions from EDGARv4.2 + LPJ



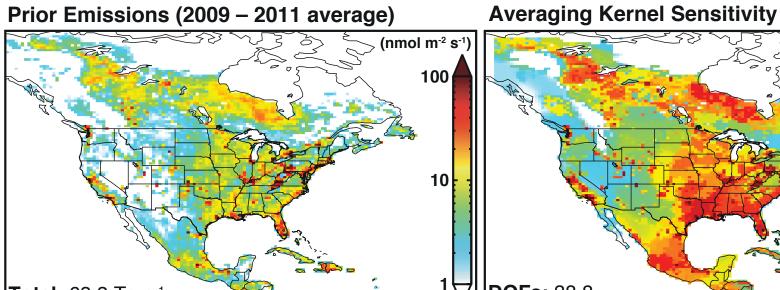


**Total:** 63/537 Tg a<sup>-1</sup>

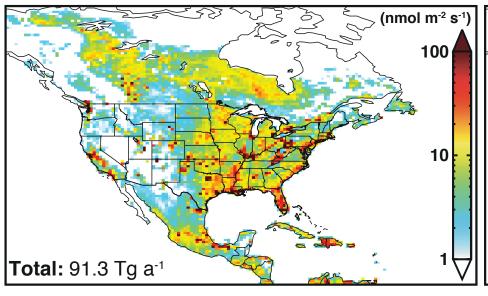
North America Global

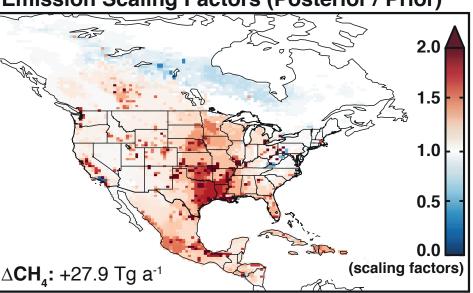


### Constraining North American methane sources









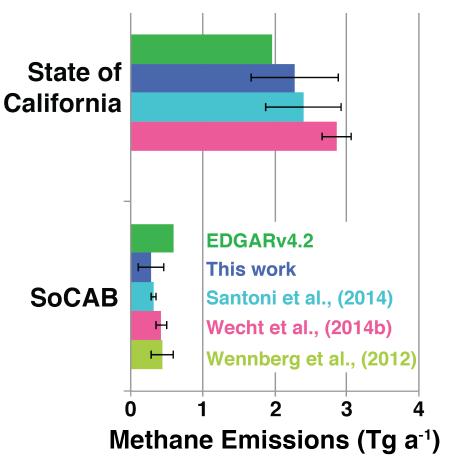
(unitless)

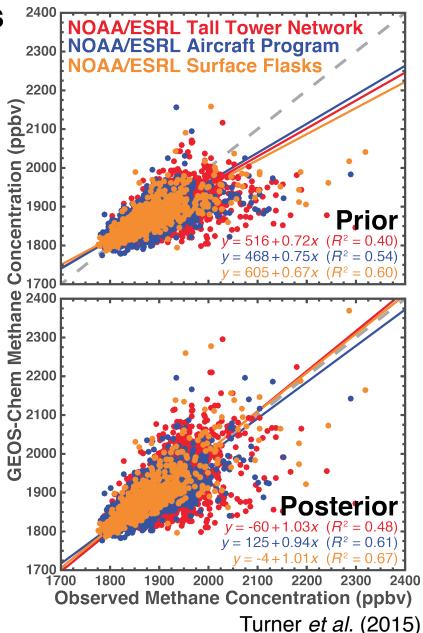
0.01

Turner *et al.* (2015)

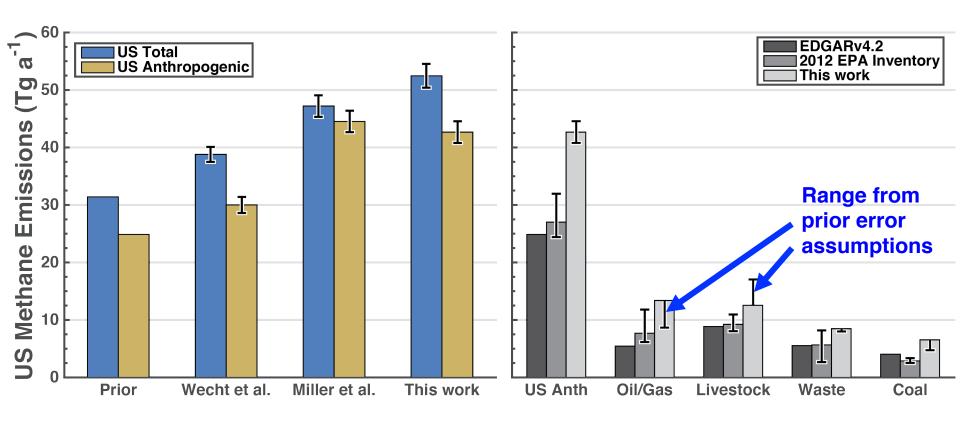
# Does this posterior inventory improve things?

- Consistent emission estimates with regional and local studies
  - Improves comparison with independent observations!



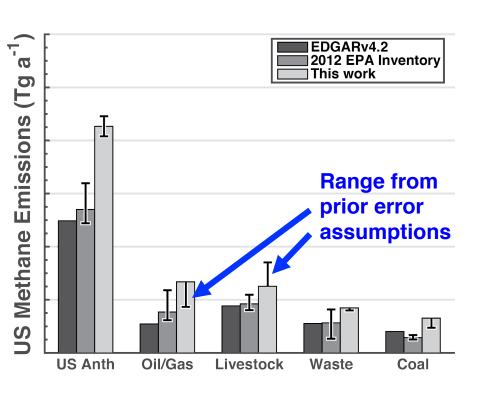


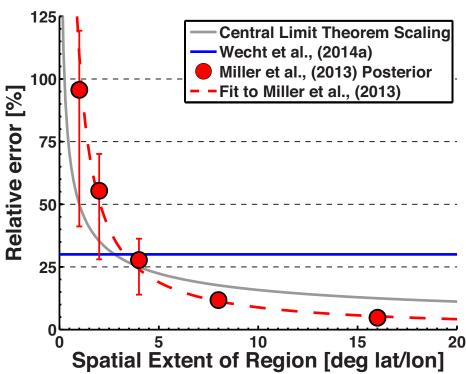
#### US methane emissions and source attribution



- ▶ US emissions are a factor of 1.5 larger than the US EPA
- Livestock + Oil/Gas are the largest underestimated sources
- Attribution is sensitive to assumption about the prior error

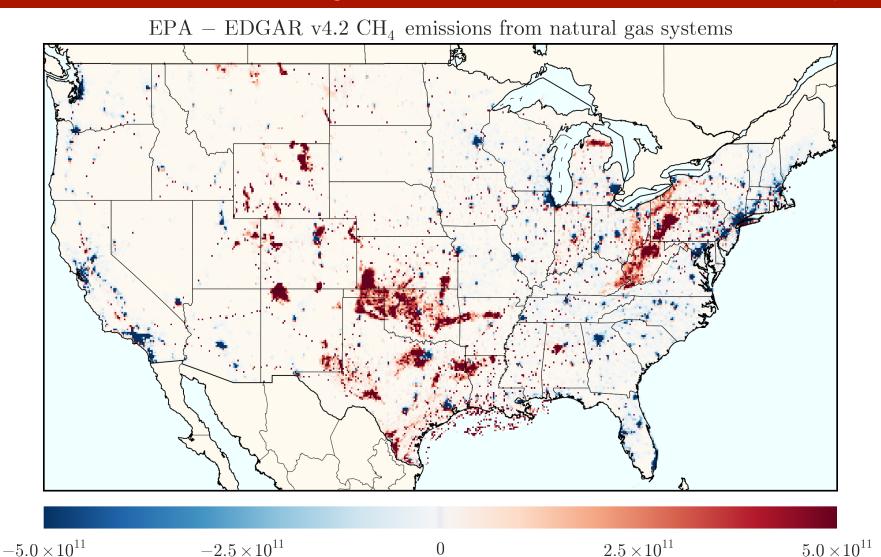
#### US methane emissions and source attribution





- ▶ Partitioning between oil/gas and livestock is dependent on specification of prior error
  - Prior error like Wecht et al. (2014a) yields more livestock emissions
  - Prior error like CLT (more similar to Miller et al.) yields balance between oil/gas and livestock

# Development of a gridded EPA methane inventory

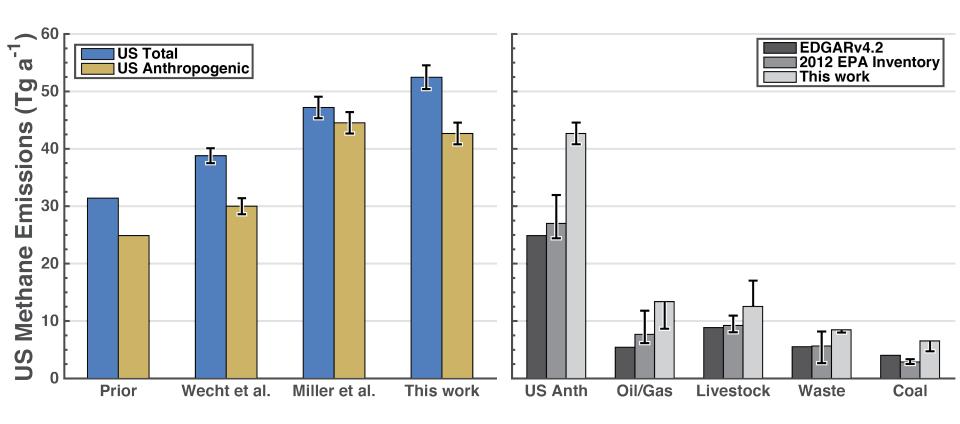


Improves potential of inversions to test and improve the EPA inventory

 $\mathrm{CH_4}\ \mathrm{emissions}\ (\mathrm{molec\ s^{-1}\ cm^{-2}})$ 

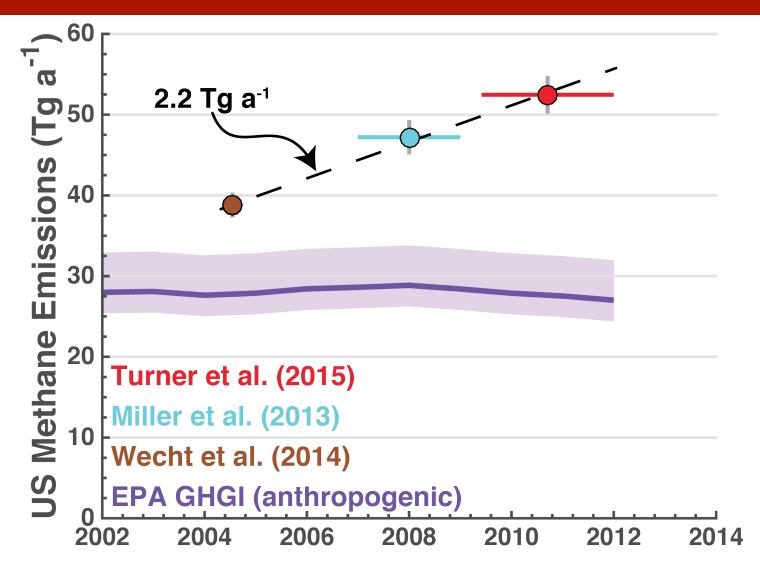
Maasakkers *et al.* (in prep)

#### Trend in US methane emissions?



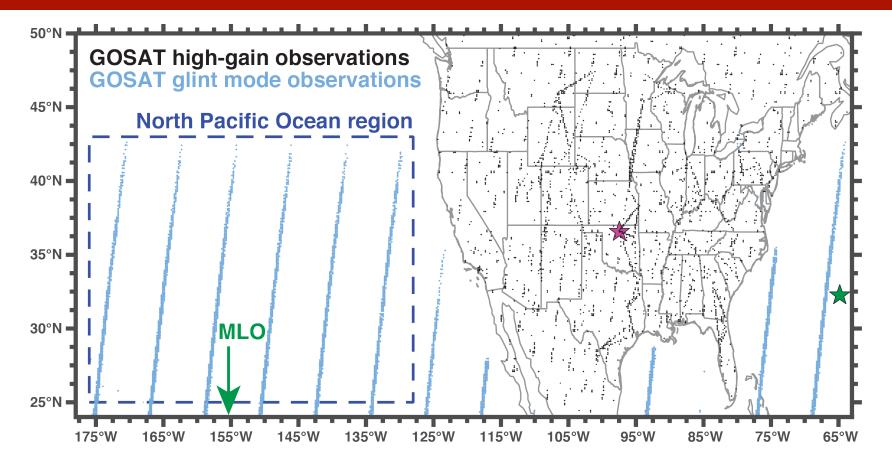
▶ What about the difference in magnitude between Wecht et al., Miller et al., and Turner et al.?

#### Trend in US methane emissions?



Top-down studies point to an increase in US methane, not seen in bottom-up estimates

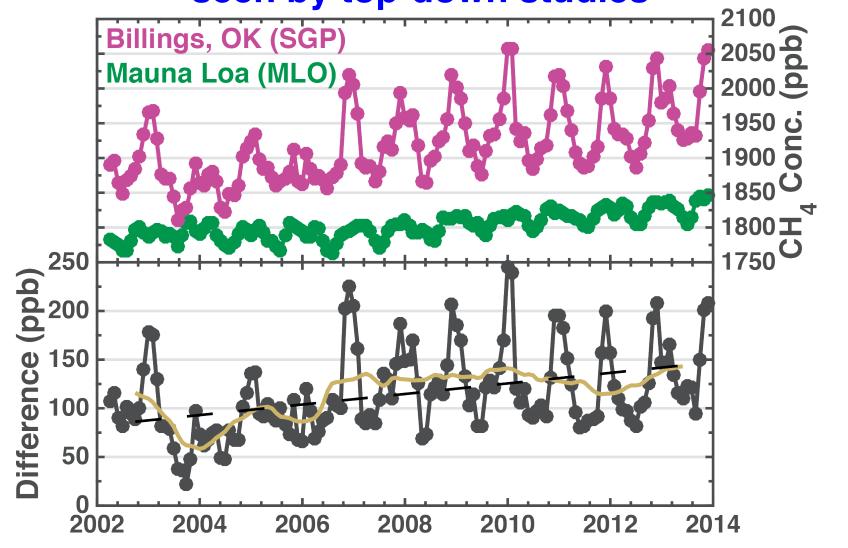
#### What data do we have to corroborate this trend?



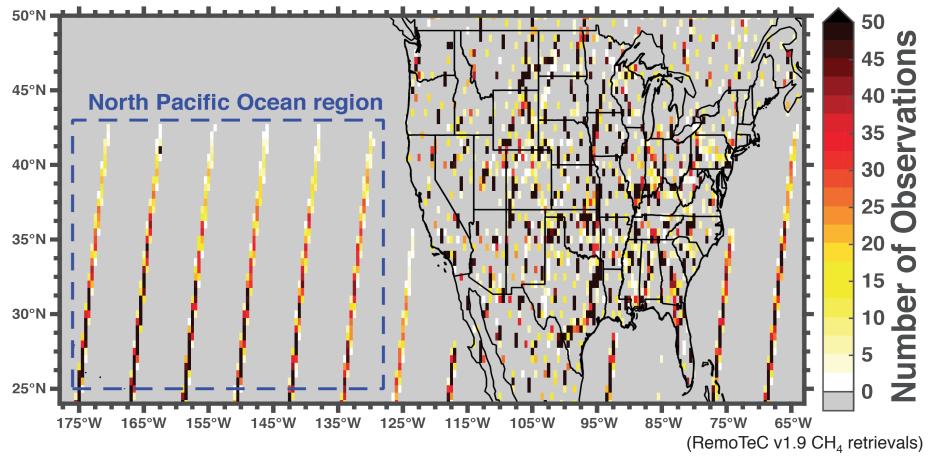
- Surface observations from the NOAA/ESRL flask network
- Nadir-mode observations from the GOSAT satellite
- Glint-mode observations from the GOSAT satellite

## Increasing difference in NOAA/DOE observations

Coincides with increase in US methane emissions seen by top-down studies

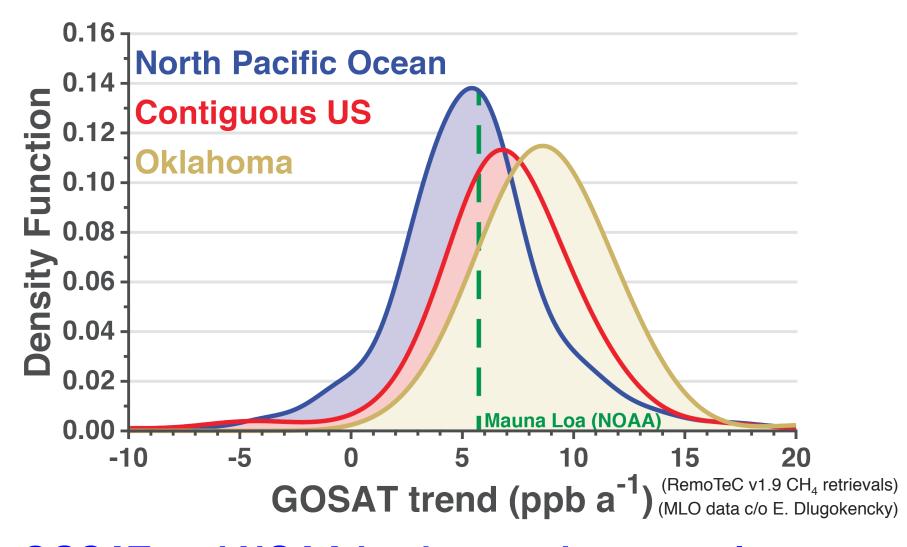


## Use GOSAT for regional trend analysis



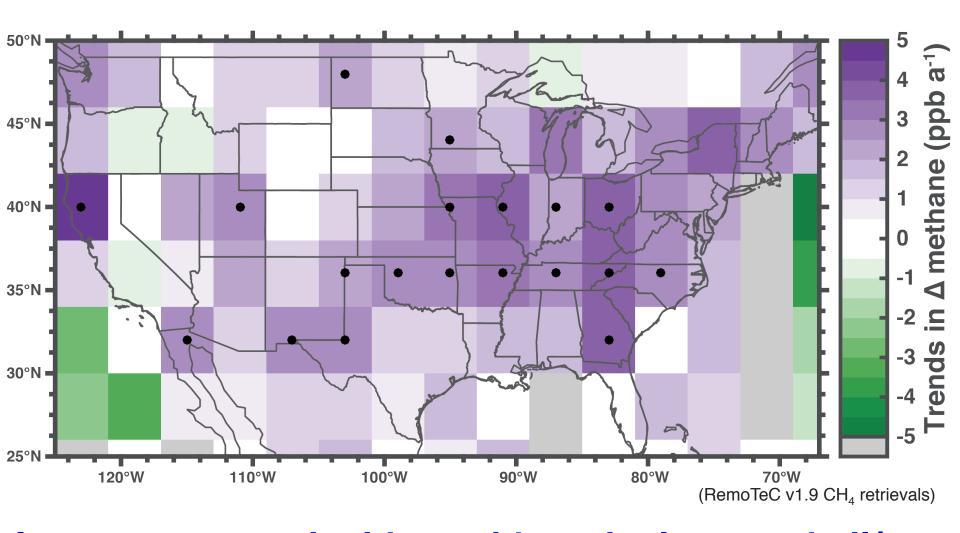
- Look at trends over locations where GOSAT samples
- Compare ocean glint to contiguous US observations

## Increasing difference in GOSAT observations



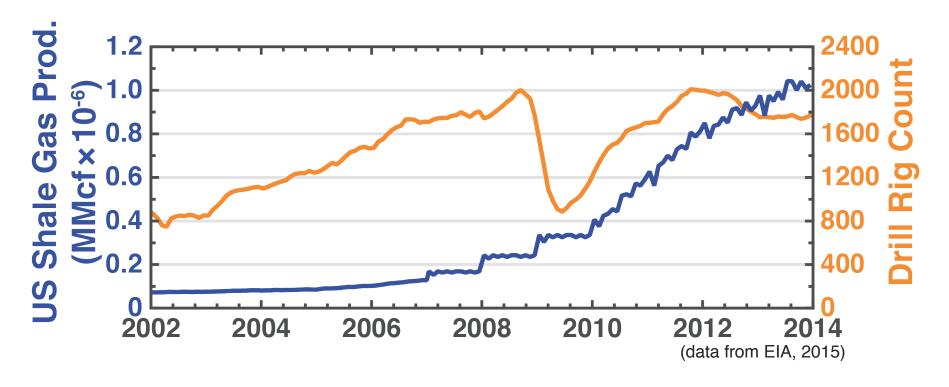
GOSAT and NOAA background are consistent Contiguous US enhanced from background

## Where do we find regional trends?



Increases are coincident with agriculture and oil/gas

#### Potential cause of the increase in US emissions



- ▶ 9-fold increase in US shale gas production from 2002–2014
- ▶ 125% increase in active drill rigs from 2002–2014

Potentially explained by oil/gas increases

## Summary

- Space-borne observations can be used to estimate regional methane emissions
- US methane emissions have increased more than 30% in the past decade
  - Likely due to anthropogenic (oil/gas or agriculture) sources
- Could be a driver in the renewed methane growth

